## **Performance Test**

Date	18 November 2022
Team ID	PNT2022TMID36549
Project Name	Project - Developing a Flight Delay Prediction Model using Machine Learning
Maximum Marks	10 Marks

## Model Performance Testing:

S.No.	Parameter	Values	Screenshot
1.	Metrics	Classification Model: Confusion Matrix - , Accuray Score- & Classification Report -	Classification Report
			<pre>print(classification_report(Y_test, Y_pred_log_test))</pre>
			precision recall f1-score support
			0.0 0.96 0.94 0.95 1985
			1.0 0.60 0.73 0.66 262
			accuracy 0.91 2247
			macro avg 0.78 0.83 0.81 2247
			weighted avg 0.92 0.91 0.92 2247
			Accuracy, Precision, Recall, F1 Score
			: acc_log = accuracy_score(Y_test, Y_pred_log_test) prec_log, rec_log, ft_log, sup_log = precision_recall_fscore_support(Y_test, Y_pred_log_test) print('Accuracy_Score =', acc_log) print('Precision =', prec_log[0]) print('Recall =', rec_log[0]) print('F1 Score =', f1_log[0])
			Accuracy Score = 0.9127725856697819 Precision = 0.9632314862765406 Recall = 0.9370277078085643 F1 Score = 0.9499489274770173
			Checking for Overfitting and Underfitting
			<pre>log_train_acc = accuracy_score(Y_train, Y_pred_log_train) log_test_acc = accuracy_score(Y_test, Y_pred_log_test) print('Training Accuracy =', log_train_acc) print('Testing Accuracy =', log_test_acc)</pre>
			Training Accuracy = 0.9205253784505788 Testing Accuracy = 0.9127725856697819

		Confusion Matrix
		<pre>pd.crosstab(Y_test.ravel(), Y_pred_log_test)</pre>
		col_0 0.0 1.0
		row_0
		0.0 1860 125
		<b>1.0</b> 71 191
Tune the Model	Hyperparameter Tuning - Validation Method -	Tuning the Hyper Parameters of Logistic Regression
		<pre>parameters = { 'solver':['newton-cg', 'lbfgs', 'liblinear'],</pre>
		In [57]: tuned_model = GridSearchCV(LogisticRegression(max_iter=800), param_grid=parameters, verbose=2) tuned_model.fit(X_train, Y_train.ravel())
		Out[57]: GridSearchCV(estimator=LogisticRegression(max_iter=800),  param_grid={'C': [100, 10, 1.0, 0.1, 0.01], 'penalty': ['12'],  'solver': ['newton-cg', 'lbfgs', 'liblinear']},  verbose=2)
		Testing the Tuned Model
		<pre>Y_pred_tun_train = tuned_model.predict(X_train) Y_pred_tun_test = tuned_model.predict(X_test)</pre>
		pd.DataFrame(Y_pred_tun_train).value_counts()
		: 0.0 7734 1.0 1250 dtype: int64
		pd.DataFrame(Y_pred_tun_test).value_counts()
		: 0.0 1922
		1.0 325 dtype: int64
	Tune the Model	

## Evaluating the Timed Model using I¥letrics Classification Report pnint(classification\_report(Y\_test, Y\_pred\_tun\_test)) precision recall f1-score support B.B 6. 97 B.94 6. 9S 1.0 0.61 B.7d 0. 6B 1985 262 accuracy macro avg 0.79 weighted avg 6.93 0.92 2247 B.BS B.92 0. 81 2247 8.92 2247 Accuracy Pzecis•n, recall, ri scoe plec ten, rec tun, f't ten, sup tun precision recall fscore support(Y test, 'Y peed tun test) print('Recall =', rec ten[8]) print( £1 Score =', of tun[e]) Accuracy Score - 0.9158878504672897 Frecision = 8.9672216441287075 Recall - 6.9J65239294710J28 FI Score = 0.95J6252879447147 cnecfi:ing ffir Overfia\*rig ang Un#erfiainp tun\_train\_acc = accuracy\_score(Y\_train, Y\_pred\_tun\_train) tun test acc = accuracy score(Y test, Y pred tun test) print('Training Accuracy =', tun\_train\_acc) pnint('Testing Accuracy =', tun test acc) Tra1 ning Accuracy = 0 . 9213B45414B694S 7 Testing Ac cureacy = B . 9158878504672897 Confusion Matrix pd.crosstab(Y test.ravel(), Y pred tun test) col 0 0.0 1.0

0.0 1859 126

63 199

1.0